

# How I Think

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# Overview

Discuss thinking as Economist vs. Physicist

- ▶ Data
- ▶ Model
- ▶ Verification & Validation
- ▶ Theory vs. Experiment/Applied

Case Study: Finite sample bias in the BLP model

# Theory vs. Experiment

Theory and experiment have a different relationship in Physics and Economics:

- ▶ In Physics:
  - ▶ Tight coupling promotes progress in both
  - ▶ Frequent falsification of models via experiment
  - ▶ Verification of theoretical predictions
- ▶ In Economics:
  - ▶ Theorists often work with toy models which are untestable
  - ▶ Theory often provides clever mathematical framework to justify our 'intuition' about real world behavior
  - ▶ Empiricists develop own statistical models

# Data : Natural Sciences

In Natural Sciences research often proceeds as follows:

- ▶ Write grant
- ▶ Build experiment
- ▶ Collect data from a controlled experiment (unless you are an Astronomer...)
- ▶ Write paper
- ▶ Repeat

# Data : Social Sciences (1)

In Social Sciences data is usually of lower quality and designed for a different purpose:

- ▶ Survey data
- ▶ Purchase marketing data (very expensive)
- ▶ Census data
- ▶ Beg or cajole data from some personal connection
- ▶ But, many confounding factors, most of which are unknown and unobserved

Economists are similar to Astronomers: you can only observe...

## Data : Social Sciences (2)

- ▶ Natural experiment: Is it truly randomized?
- ▶ Sometimes run a controlled experiment:
  - ▶ Field experiment
  - ▶ In the laboratory
  - ▶ Undergraduates most common test subjects
  - ▶ Dependent on Human Subjects Committee
  - ▶ Does it have external validity?

Poor funding to collect good data to answer important social questions even when real money is at stake!

## Data : Social Sciences (3)

Endogeneity – unobserved shocks can affect both LHS and RHS:

1. Measurement error
  2. Omitted Variable Bias
  3. Simultaneity : an unobserved shock which affects RHS and LHS variables.
- ▶ Physics concerned with minimizing 1.
  - ▶ But, in Economics 2. and 3. often dominate

### Example

What if  $\epsilon_{it}$  contains unobserved Ability?

$$Wage_{it} = \alpha School_{it} + \beta Class_i + \gamma FamilyWealth_{it} + \epsilon_{it}$$

# Model Building : Physics

Models are usually governed by tight feedback loop with the real world:<sup>1</sup>

- ▶ Iterative refinement to understand one set of laws
- ▶ Some ever improving approximation to the truth is knowable
- ▶ Driven by explaining observations from natural world
- ▶ Everyone is basically refining the same model
- ▶ General principles/laws apply to all disciplines, e.g. conservation of energy
- ▶ Knowledge is deep like a pine tree

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<sup>1</sup>Let's not talk about string theory...



# Model Building : Economics (1)

In Economics, model building is part art and part science:

- ▶ The world is complex
- ▶ We are ignorant
- ▶ Many (important) factors are unknowable

Good Economics handles this by

- ▶ Choosing a sensible functional form to capture relationships of interest
- ▶ Using an error structure which soaks up confounding factors
- ▶ Identifying important factors while minimizing potential sources of bias

## Model Building : Economics (2)

There are economic 'laws' but many depend on the specific problem and application:

- ▶ A good economist is conversant in a much larger number of models
- ▶ Models often have narrow applicability
- ▶ Some cannot be tested
- ▶ 'I did not write this model to be tested. I wrote it to tell a story.'
- ▶ Knowledge is more like a dense hedge

# Verification

Economics lacks a culture of Verification:

- ▶ No credit for replication studies
- ▶ Consequently, results are almost never verified
- ▶ Huge push to publish single author papers to ‘signal’ your ability, especially early in career
- ▶ Sloppy numerical work in most cases:
  - ▶ Code and data are unobtainable
  - ▶ No unit tests
  - ▶ Few Monte Carlo studies to verify
  - ▶ Backlash for discussing or asking about numerical details
- ▶ When we mentioned we were going to test our estimation code on Monte Carlo data, a colleague said, ‘Oh, you are serious. . .’

# Validation

Some concern about validation:

- ▶ Testing for model misspecification
- ▶ Usually consists of nesting model in a larger statistical model
- ▶ No scenario analysis
- ▶ No attempt to explore unknown unknowns
  - ▶ Focus is on factors causing unobservable shocks
  - ▶ Little attempt to determine magnitudes of different unobservables
- ▶ Avoidance of anything outside standard toolbox: e.g. preference for pMC over SGI quadrature rules

## Case Study: BLP

The Berry, Levinsohn, and Pakes (1995) model of differentiated products dominates the field of Industrial Organization:

- ▶ Justification of model rests on one paper
  - ▶ Only considers a special case
  - ▶ Only uses tools in mainstream Economics toolbox
- ▶ Drives  $\sim 70\%$  of papers in applied Industrial Organization
- ▶ Will soon be used for anti-trust inquiries. . .

# 'The Death Star'

Skrainka (2011) develops an infrastructure to test BLP and other econometric models:

- ▶ Prior simulation experiments only look at data with one market
- ▶ First simulation study of bias for range of (more) realistic datasets
- ▶ Rigorous generation of synthetic data as well as estimation: most econometric theorists use antediluvian numerical methods
- ▶ Automated everything possible to avoid stupid errors
- ▶ Developed scripts (bash, Python, R) to verify each run
- ▶ One of the largest simulations in Economics – but only about 90,000 CPU-hours



# Data Generation

Generating realistic data is harder than estimating the model:

- ▶ Used structural model so data has correct statistical properties as number of markets,  $T$ , and products,  $J$ ,  $\rightarrow \infty$
- ▶ Smaller datasets are subsets of larger datasets
- ▶ Generated prices from Bertrand-Nash equilibrium:
  - ▶ Must find root of nonlinear set of equations which often mislead a solver
  - ▶ Has  $T$  dense  $J \times J$  blocks
  - ▶ Used Path 5.0.00
- ▶ Similar technology to estimation (Eigen, C++, etc.)
- ▶ Generated up to 50 markets and 100 products

# Data Estimation

Used current best methods for estimation

- ▶ Su & Judd (2011): Bi-level optimization + SNOPT solver
- ▶ Skrainka & Judd (2011): Sparse grids quadrature rules
- ▶ Eigen, C++, higher precision (long double)
- ▶ 50 starts to hopefully find global min
- ▶ Restarts to ensure found global min
- ▶ Used two different instrumentation strategies



# Parallelization

Very easy to parallelize:

- ▶ Each replication and start is independent
- ▶ Use Parameter Sweep
- ▶ Very easy with PBS
- ▶ Compute results based on best start

## Bias: Point Estimates

T	J	Bias	Mean Abs Dev	Med Abs Dev	RMSE
1	12	-2	3	1.3	5.7
1	24	-0.72	1.9	1.2	3.2
1	48	-0.52	1.9	1.2	3
1	100	-0.57	1.7	1.3	2.3
10	12	-1.7	2.6	1.1	6
10	24	-0.65	2	1.3	3.6
10	48	-0.64	1.9	1.3	3.2
10	100	-0.83	2	1	3.9
25	12	-0.62	1.9	1.2	3.1
25	24	-0.96	2.3	1.4	3.7
25	48	-1.3	2.8	1.2	7.6
25	100	-0.95	2.1	1.1	3.7
50	12	-0.39	1.6	1.1	2.7
50	24	-1.2	2.5	1.1	5.4
50	48	-1.2	2.2	1.3	5.2
50	100	-0.63	1.9	1.3	3

## Bias: Own-Price Elasticities

T	J	Bias	Mean Abs Dev	Med Abs Dev	RMSE
1	12	-0.77	2.2	0.94	4.9
1	24	-0.095	1.5	0.77	3.3
1	48	-0.082	1.6	0.91	2.7
1	100	-0.39	1.5	0.98	2.5
10	12	-0.5	1.7	0.81	3.3
10	24	-0.57	1.7	0.83	3.3
10	48	-0.16	1.5	0.97	2.2
10	100	-0.53	1.7	0.93	3.3
25	12	-0.3	1.4	0.94	2.7
25	24	-0.72	1.8	1.1	3
25	48	-0.87	2.2	1.1	4.9
25	100	-0.61	1.7	0.97	2.7
50	12	-0.43	1.5	0.94	2.6
50	24	-0.77	1.9	0.91	3.8
50	48	-0.97	1.9	1.1	4
50	100	-0.56	1.8	1.1	2.9

## Bias: Quadrature

	Bias		Mean Abs Dev		Med Abs Dev		RMSE	
	SGI	pMC	SGI	pMC	SGI	pMC	SGI	pMC
$\theta_{11}$	0.96	12.34	2.29	13.25	1.20	3.64	4.00	28.92
$\theta_{12}$	0.02	-0.13	0.52	0.38	0.22	0.33	0.94	0.48
$\theta_{13}$	-0.28	-0.38	1.47	1.21	0.62	0.99	3.01	1.51
$\theta_{21}$	22.57	128.22	23.01	128.24	2.62	34.06	81.76	253.87
$\theta_{22}$	0.02	-0.04	0.12	0.16	0.07	0.13	0.19	0.20
$\theta_{23}$	0.08	0.64	0.36	0.75	0.16	0.79	0.75	0.90

**Table:** Comparison of bias in point estimates : SGI vs. pMC for T=2 markets and J=24 products with 165 nodes.

# Impact on Policy

This study will hopefully impact Econometrics and applied work driving policy:

- ▶ Asymptotic inference is not valid with this model
- ▶ Possible to test Econometric tools/estimators more rigorously than previously attempted
- ▶ BLP is poised to become a key model for anti-trust investigation but estimation strategy is a failure:
  - ▶ Traditional strategy is extremely biased
  - ▶ Best strategy is also biased!
  - ▶ Lack of control for endogeneity for typical datasets
- ▶ Conventional wisdom that increasing the number of markets will improve results is wrong because most variation is with-in market!

# Conclusion

Economists can do much more verification and validation:

- ▶ Modern resources make it easy to run large scale Monte Carlo experiments:
  - ▶ Test estimators
  - ▶ Investigate more scenarios and counter-factual policy experiments
  - ▶ Richer policy analysis
- ▶ BLP model is seriously biased:
  - ▶ Need a new model
  - ▶ Need better identification strategy
- ▶ Other models probably have similar problems, but no one has attempted to validate them
- ▶ Better data and methods needed to provide better answers to important questions